

Advances in Adaptive Control Methods

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Integrated Resilient
Aircraft Control

Constraint-Based Adaptive Control - Optimal Control Modification

Objective

- Introduces notion of **constraint-based adaptive control** that combines adaptive control with optimal control to achieve constrained error minimization.
- Develops robust **optimal control modification** adaptive law that **enforces linear quadratic constraints**.

Technical Challenges

- Persistent excitation (PE)** can adversely affect robustness of adaptive control due to high-frequency input signals.
- Nonlinear input-output mapping** of adaptive control can result in unpredictable performance.

Technical Approach

- Minimize LQ cost function $J = \lim_{t \rightarrow \infty} \frac{1}{2} \int_0^t [e(t) - \Delta(t)]^T Q [e(t) - \Delta(t)] dt$ subject to error dynamics $\dot{e}(t) = A_m e(t) + B [\tilde{\Theta}^T(t) \Phi(x(t)) - \epsilon(x(t))]$
- Approach based on application of Pontryagin's Minimum Principle
- Optimal Control Modification Adaptive Law**

$$\dot{\Theta}(t) = -\Gamma \Phi(x(t)) [e^T(t) P - \nu \Phi^T(x(t)) \Theta(t) B^T P A_m^{-1}] B$$

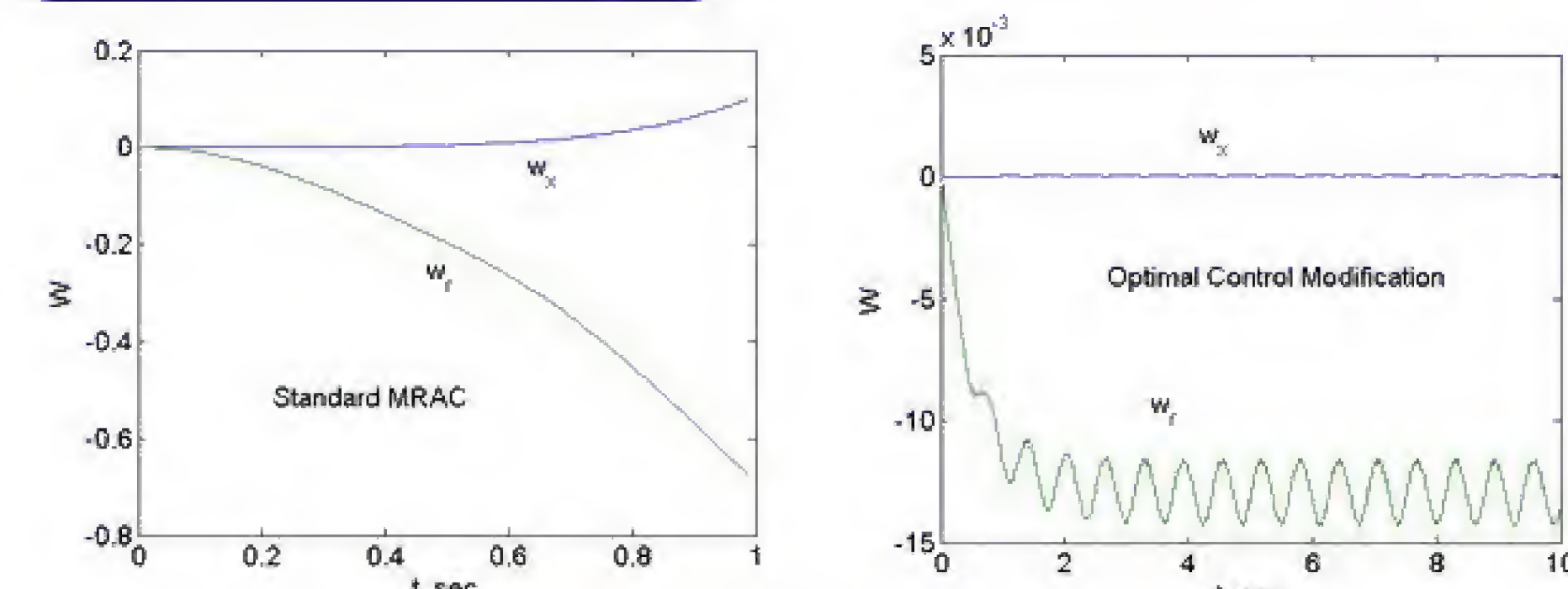
Lyapunov stability proof shows that the adaptive law is stable and tracking error is UUB.

- Modification term proportional to persistent excitation (PE)** to counteract adverse effects of PE

Example

$$\begin{aligned} \dot{x} &= -x + 2u - 0.1y \\ \ddot{y} + 10\dot{y} + 100y &= 7x \\ r &= 1 + 0.1 \sin 10t \\ u &= -0.5x + r - w_x x - w_r r \end{aligned}$$

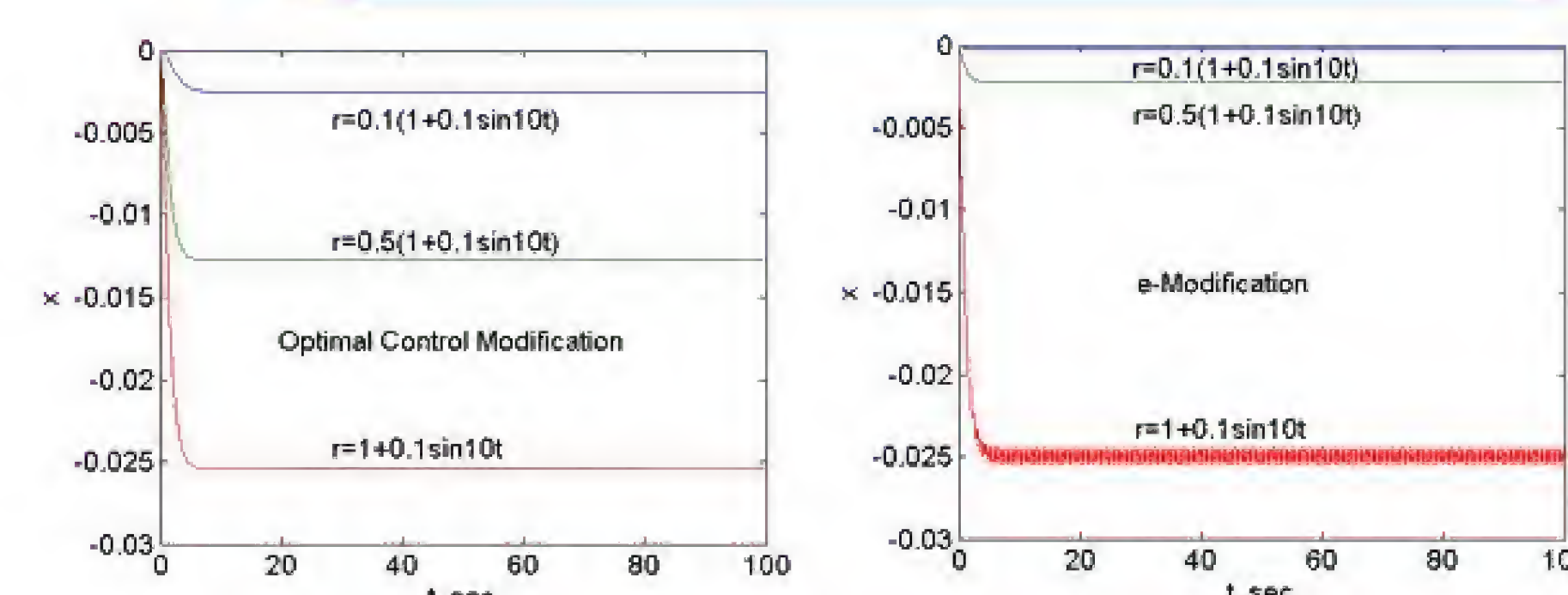
1st-order plant with 2nd-order unmodeled dynamics and input at the same frequency as that of unmodeled dynamics



Robustness to Unmodeled Dynamics

Asymptotic Linearity for Linear Uncertainty

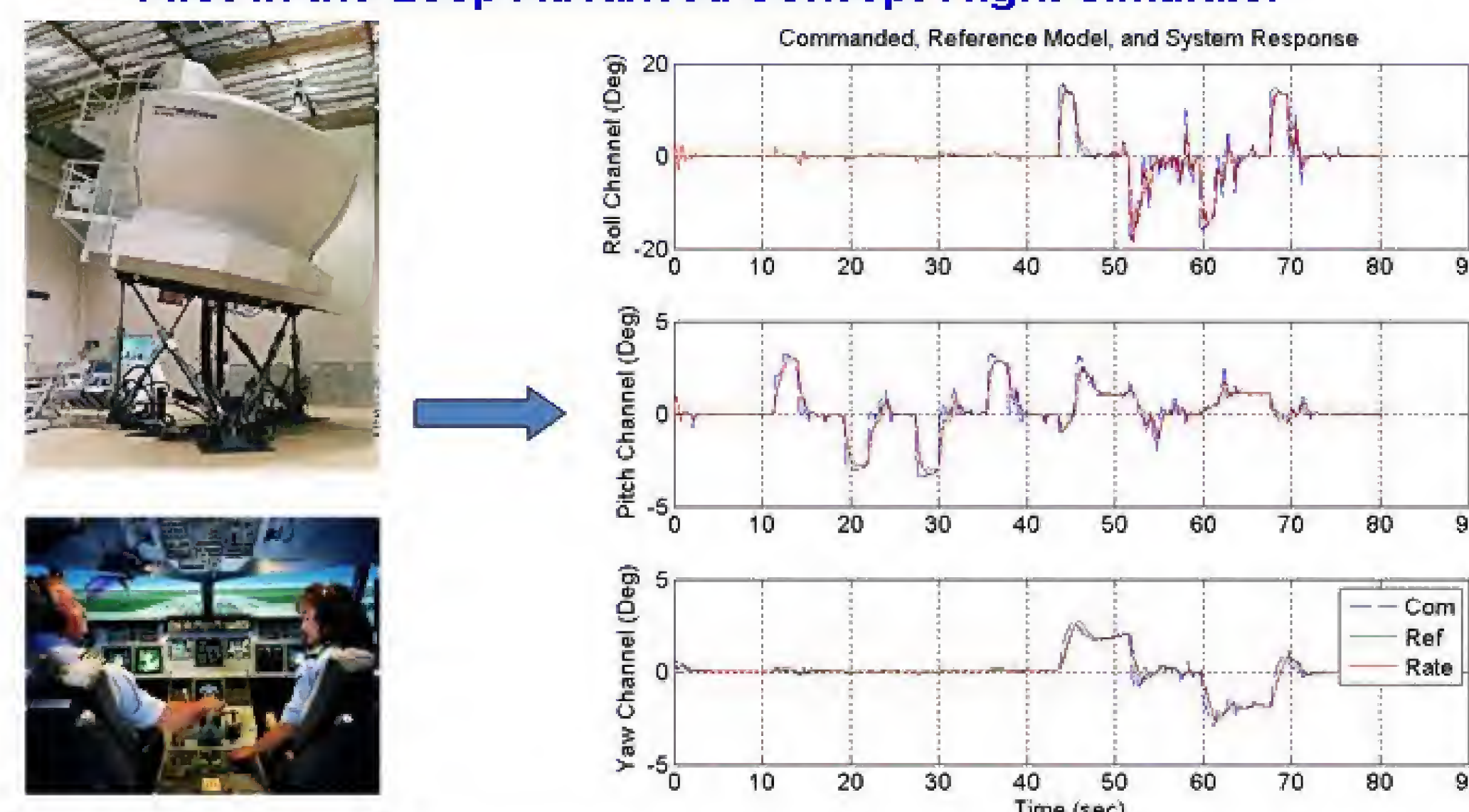
- Fast adaptation condition $\Phi^T(x(t)) \Gamma \Phi(x(t)) \gg \|A_m\| \gg 1$
- Asymptotic behavior $B \Theta^T(t) \Phi(x(t)) \rightarrow \frac{1}{\nu} P^{-1} A_m^T P e(t)$
- Linear tracking error dynamics for linear uncertainty $\dot{e}(t) = -P^{-1} \left[\left(\frac{1+\nu}{2\nu} \right) Q - \left(\frac{1-\nu}{2\nu} \right) S \right] e(t) - B \Theta^{*T} x(t)$
- Adaptive law can be designed to guarantee stability** for given bound on Θ^* using projection operator such that $A_c = -P^{-1} \left[\left(\frac{1+\nu}{2\nu} \right) Q - \left(\frac{1-\nu}{2\nu} \right) S \right] + B \Theta^{*T}$ is Hurwitz and satisfies linear stability margin requirements everywhere inside projection bound \Rightarrow **certifiable adaptive control**
- Note: e-modification or sigma-modification results in nonlinear error dynamics even for linear uncertainty



Asymptotic Input-Output Linear Mapping

Pilot-in-the-Loop GTM Simulations

Pilot-in-the-Loop Advanced Concept Flight Simulator



10K ft, 250 Kn IAS
A = 0, B scaled
Doublet to capture flight director task

Summary

- Optimal Control Modification can provide stable fast adaptation to improve tracking
- Asymptotic linearity with fast adaptation can guarantee linear stability for linear structured uncertainty
- Pilot-in-the-loop simulations demonstrate effectiveness of the method

Adaptive Control of Time-Delay Systems - Time Delay Margin of MRAC

Objective

- Develops **stability analysis** for **time-delay adaptive system** and analytical tool to compute **time delay margin** (TDM) based on **Bounded Linear Stability Analysis**

Technical Challenges

- Currently no analytical tool exists to provide non-conservative and practical TDM estimate.

Technical Approach

- Input-delay adaptive system**

$$\begin{aligned} \dot{x}(t) &= Ax(t) + B[u(t - t_d) + \Theta^{*T} \Phi(x(t))] & u_{ad}(t) &= \Theta^T(t) \Phi(x) \\ u(t) &= K_x x(t) + K_r r(t) - u_{ad}(t) & \dot{\Theta}(t) &= -\Gamma \Phi(x(t)) e^T(t) P B \end{aligned}$$

- Bounded Linearity Stability** approximates adaptive system as a locally bounded LTI system using time-window analysis

$$\begin{aligned} \dot{\Theta}^T(t) \Phi(x(t)) &= -B^T P e(t) \Phi^T(x(t)) \Gamma \Phi(x(t)) \approx -\gamma B^T P e(t) \\ \gamma &= \frac{1}{T_0} \int_{t_i}^{t_i+T_0} \Phi^T(x(\tau)) \Gamma \Phi(x(\tau)) d\tau > 0 \in \mathbb{R} \\ t &\in [t_i, t_i + T_0] \quad i = 0, 1, \dots, n \end{aligned}$$

$$\dot{u}_{ad}(t) \approx -\gamma B^T P e(t) + \Theta^T(t) \dot{\Phi}(x(t))$$

γ can be viewed as integral product of adaptive gain and PE value

- Locally LTI approximation of tracking error dynamics

$$\begin{bmatrix} \dot{e}(t) \\ e(t) \end{bmatrix} = C_i \begin{bmatrix} \dot{e}(t) \\ e(t) \end{bmatrix} - D_i \begin{bmatrix} \dot{e}(t - t_d) \\ e(t - t_d) \end{bmatrix} + \begin{bmatrix} d_1(t) + d_2(t - t_d) + d_3 \\ 0 \end{bmatrix}$$

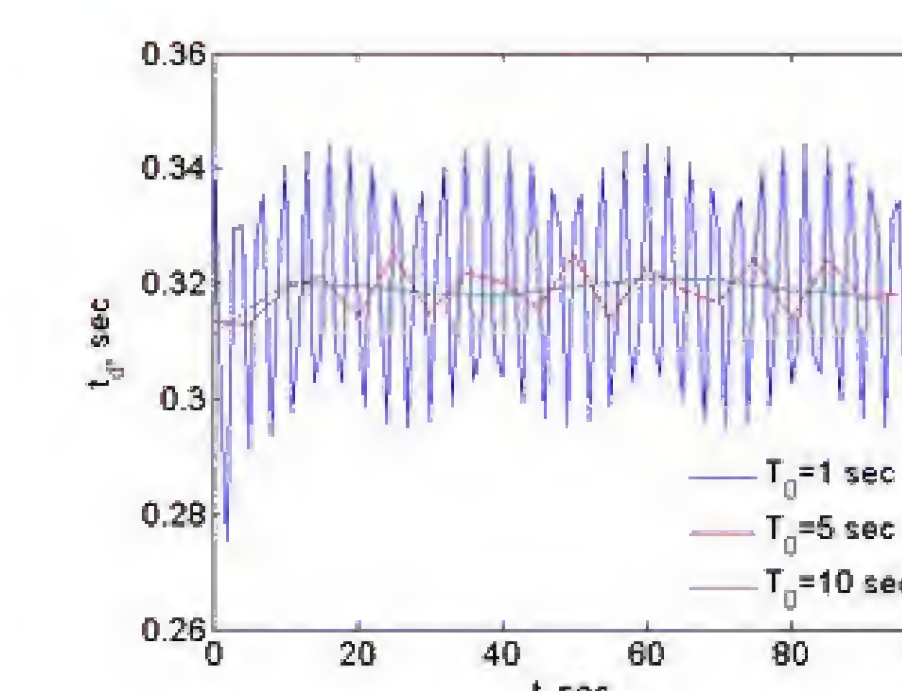
$$C_i = \begin{bmatrix} A + B \Theta^{*T} \Phi'_i & 0 \\ I & 0 \end{bmatrix} \quad D_i = \begin{bmatrix} A - A_m + B \Theta_i^T \Phi'_i & B B^T P \\ 0 & 0 \end{bmatrix}$$

- TDM estimation by matrix measure approach** - system is locally stable if time delay is less than TDM

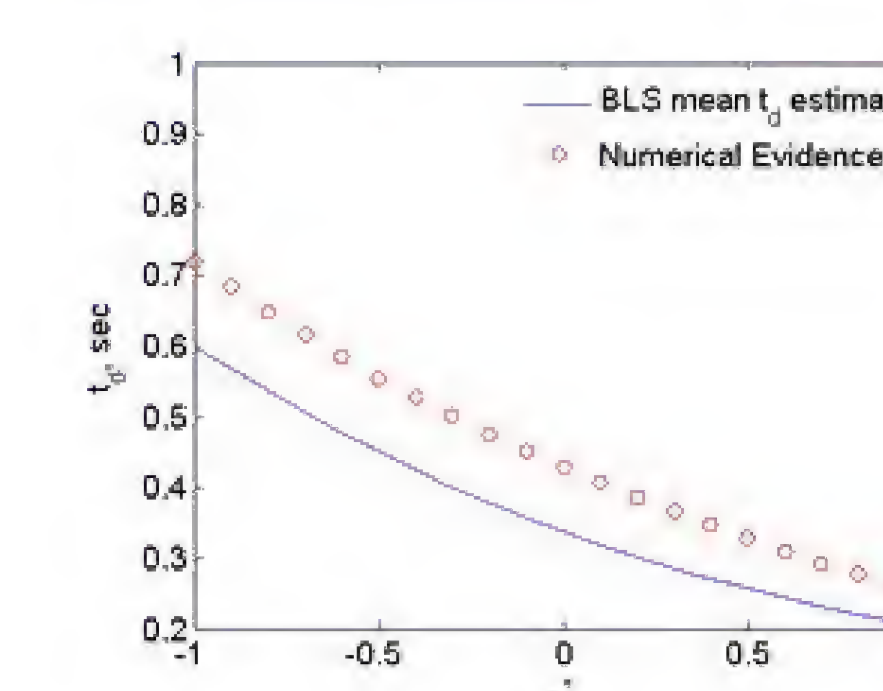
$$\begin{aligned} \omega_i &< \bar{\mu}(-jC_i) + \|D_i\| \\ t_{di} &< \frac{1}{\omega_i} \cos^{-1} \frac{\bar{\mu}(C_i) + \bar{\mu}(jD_i)}{\|D_i\|} \end{aligned}$$

Example

$$\begin{aligned} \dot{x} &= x + u + \theta^* x \\ \dot{x}_m &= -x_m + \sin t \\ u &= -2x + \sin t - \theta x \end{aligned}$$



TDM estimate agrees well with simulation results



Summary

- New analytical method provides non-conservative TDM estimate
- Method can easily be extended to sigma-modification and optimal control modification

Matrix Measure Properties

$$\bar{\mu}(C) = \max_{1 \leq i \leq n} \lambda_i \left(\frac{C + C^*}{2} \right) = \lim_{\epsilon \rightarrow 0} \frac{\|I + \epsilon C\| - 1}{\epsilon}$$

$$\underline{\mu}(C) \leq \text{Re} \lambda_i(C) \leq \bar{\mu}(C) \quad \text{Im} \lambda(C) \leq \bar{\mu}(-jC)$$

$$\bar{\mu}(C) \leq \|C\|$$

Given $\dot{x}(t) = Ax(t) - BKx(t - t_d)$, $\lambda(A - BK) \in \mathbb{C}^-$

$$\text{System is stable if } t_d < \frac{1}{\omega} \cos^{-1} \frac{\bar{\mu}(A) + \bar{\mu}(jBK)}{\|BK\|}$$

System is stable independent of time delay if

$$\bar{\mu}(A) < \|BK\| < -\underline{\mu}(A)$$